

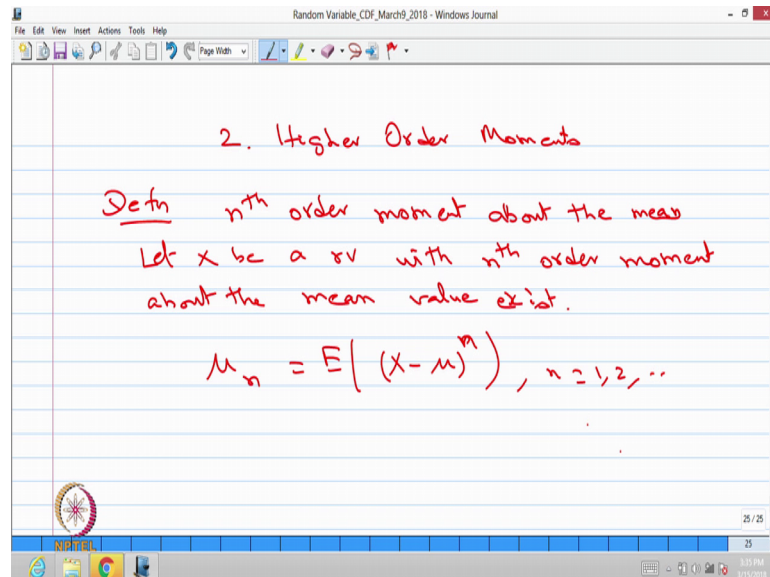
Introduction to Probability Theory and Stochastic Processes
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Lecture – 14

In this week we started Moments and Inequalities. Already we discussed in the last lecture mean and variance, mean is nothing but the first order moment and variance is nothing but the second order moment of the random variable. In the last class we have discussed the first order moment and the second order moment with the examples. In this lecture we are going to discuss higher order moments.

Since we have already discussed first and second order moment now we are going to discuss any n th order moment for the random variable if it exist followed by we are going to discuss the moments inequalities. So, let me start with definition of higher order moments, higher order moments.

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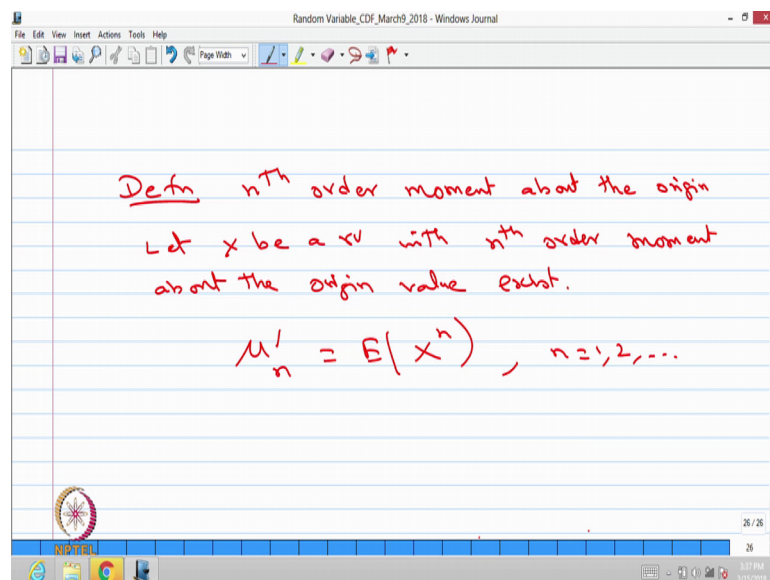


The definition that is a n th order moment about the mean. Let X be a random variable with n th order moment about the mean value exist. Then one can define with the notation μ_n suffix n that is nothing but expectation of X minus, the expectation of X is denoted by μ that is mean X minus μ power n that is going to be the n th order moment about the mean. Whenever it exist it can denoted by μ_n suffix n , whenever it exist that is the

right hand side expectation exist then you can denote by the letter mu suffix n that is expectation of X minus mu power n, where n can takes the value it could be 1 2 and so on.

Obviously, if you take the value n is equal to 1 that is nothing, but the mu suffix 1 is expectation of X minus mu that is same as expectation of X minus mu that is going to be 0, when n is equal to 2, then it is nothing but the variance of the random variable X . So, provided the right hand side expectation exist then one can define the nth order moment about the mean with the dot notation mu suffix n.

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The same way I can define the nth order moment about the origin nth order moment about the origin or some books they use a word 0, both are one of the same.

Let X be a random variable with nth order moment about the origin value exist. Then one can define with the notation mu suffix n dash that is nothing, but expectation of X power n here again n can take the value 1, 2 and so on. So, when n is equal to 1 this is nothing but the mean or expectation of the random variable and two onwards it is going to be called as a nth order moment about the origin, provided the expectation exist that is very important.

One can relate the second order moment about the origin with the second order moment about the mean.

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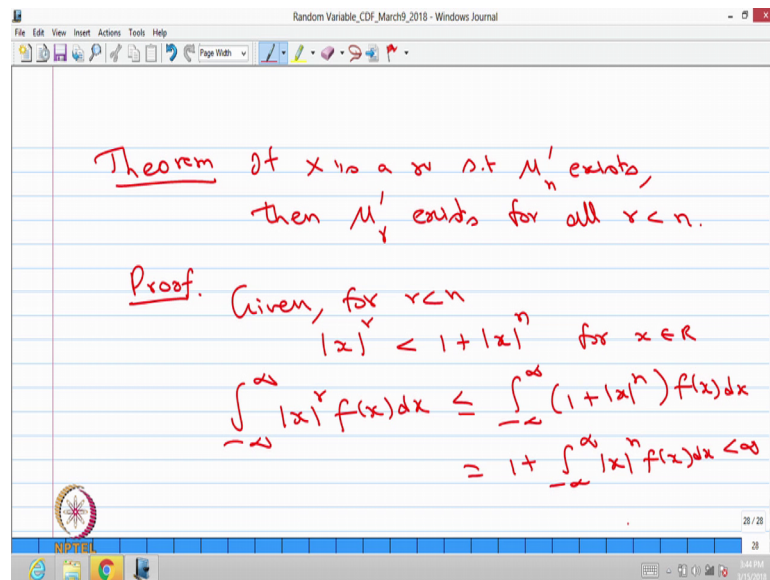
The image shows a screenshot of a software window titled "Random Variable_CDF_March9_2018 - Windows Journal". The window contains handwritten mathematical derivations in red ink on a blue-lined background. The derivations are as follows:

$$\begin{aligned}\mu_2' &= E(X^2) \\ \text{Var}(X) = \mu_2 &= E((X - \mu)^2) \\ &= E(X^2 - 2X\mu + \mu^2) \\ &= E(X^2) - 2\mu E(X) + \mu^2 \\ &= E(X^2) - 2\mu^2 + \mu^2 \\ &= E(X^2) - \mu^2 = \mu_2' - (\mu_1')^2\end{aligned}$$

For example, μ_2' that is expectation of X square and μ_2 that is nothing but the expectation of X minus μ the whole square. This is same as variance of X . So, the expectation of X minus μ whole square if you expand that is expectation of X square minus $2X\mu$ plus μ square, expectation is a linear operator so it is expectation of X square minus 2μ ; and μ are so it is expectation of X . And μ is a constant, so μ square constant expectation of μ square that is μ square

So, when you simplify you will get expectation of X square minus this is 2μ into μ therefore, 2μ square plus μ square. So, that is same as expectation of X square minus μ square that is same as μ_2' minus $(\mu_1')^2$; that means, μ_2 is nothing but μ_2' minus $(\mu_1')^2$ the whole square so that means, one can write central moment about the mean in terms of central moment about the origin.

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Next I am going to give the one important result as a theorem. What the theorem says if X is a random variable such that μ_n' exists then μ_r' exists for all $r < n$ that is a theorem.

Whenever for a random variable if the n th order moment about origin exist, then all the r th order moment about the origin exist for all $r < n$. You can give the proof of this theorem, given $|x|^r < 1 + |x|^n$ for all $x \in \mathbb{R}$, this is for all x belonging to real.

We can conclude suppose you consider X as a continuous type random variable, $\int_{-\infty}^{\infty} |x|^r f(x) dx$, where $f(x)$ is the probability density function of a continuous type random variable that is less than or equal to $\int_{-\infty}^{\infty} (1 + |x|^n) f(x) dx$ that is same as $1 + \int_{-\infty}^{\infty} |x|^n f(x) dx$. And since n th order moment about the origin exist therefore, this quantity is going to be finite. This implies $\int_{-\infty}^{\infty} |x|^r f(x) dx$ is a finite that is for all r which is less than n .

So, this is given you can include one more statement for $r < n$, given for $r < n$ $|x|^r < 1 + |x|^n$ this is true therefore, both side you can do the integration by multiplying $f(x)$ and given that its n th order moment about the origin exist therefore, for all $r < n$ the moment of r exist also.

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Theorem. Let X be a r.v. whose n^{th} order moments exist. Then

$$\mu_n = \sum_{k=0}^n \binom{n}{k} \mu_k' (-\mu_1')^{n-k}$$

Proof

$$\begin{aligned} \mu_n &= E[(X - \mu)^n] \\ &= E[(X - \mu_1')^n] \end{aligned}$$

The next result as a theorem let X be a random variable whose n^{th} order moments exist. Then one can write μ_n as a summation over k is equal to 0 to n , $\binom{n}{k} \mu_k'$ with the minus μ_1' dash power $n - k$. This can be proved whenever n^{th} order moment exist, then one can write the n^{th} order moment about the mean is same as a function of all the previous order moments about the origin.

The proof is as follows you start with μ_n that is nothing but n^{th} order moment about the mean that is $X - \mu$ power n , that is same as the expectation of $X - \mu_1'$ can write μ as a μ_1' dash that power n . Now, you can go for the binomial expansion of $X - \mu_1'$ power n , that is same as expectation of summation k is equal to 0 to n , $\binom{n}{k} X^k$ and minus μ_1' dash power $n - k$ that is same as the $\binom{n}{k}$ is a constant that is not a random μ_1' dash that is also not random therefore, expectation can be taken inside that is summation k is equal to 0 to n , $\binom{n}{k} \mu_k'$ expectation of X^k minus μ_1' dash power $n - k$.

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The image shows a screenshot of a Windows Journal window titled "Random Variable_CDF_March9_2018 - Windows Journal". The window contains handwritten mathematical equations in red ink on a lined background. The equations are:

$$\begin{aligned}\mu_n &= E\left(\sum_{k=0}^n \binom{n}{k} x^k (-\mu_1')^{n-k}\right) \\ &= \sum_{k=0}^n \binom{n}{k} E(x^k) (-\mu_1')^{n-k} \\ &= \sum_{k=0}^n \binom{n}{k} \mu_k' (-\mu_1')^{n-k}\end{aligned}$$

The window's taskbar at the bottom shows the Start button, Internet Explorer, Firefox, and Chrome icons. The system tray on the right indicates the time is 3:58 PM on 1/15/2015.

I can rewrite expectation X power k as the k th order moment about origin therefore, this is going to be summation k is equal to 0 to n , n c k μ suffix k dash multiplied by minus μ suffix 1 dash power n minus k . Because of the previous theorem when the n th order moment about mean exist that means, all the previous order also exist therefore, this is a valid statement. With the help of previous moments about the mean you can always find the moment of n th order about the origin.

In conclusion with the previous starting from first to n th order moment about the origin one can get n th order moment about the mean, one can go for one easy example of how to find the n th order moment for some random variable which is of the continuous type.

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Example 1. Let X be a continuous type rv with pdf

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}, \quad -\infty < x < \infty$$

$-\infty < \mu < \infty$
 $\sigma > 0$

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx = \mu$$

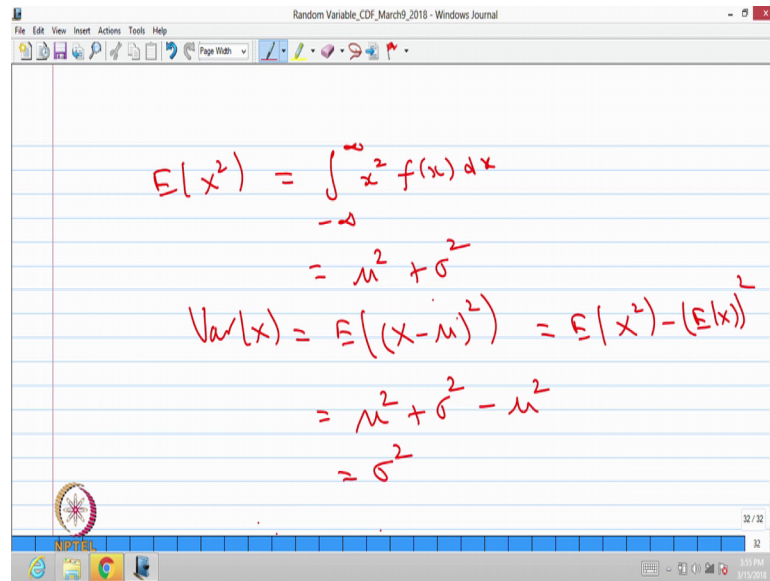
Mean of the rv X .

Let X be a continuous type random variable with probability density function f of x is 1 divided by sigma times root pi e power minus x minus μ divided by sigma the whole square multiplied by 1 by 2, where x lies between minus infinity to infinity. So, this is a probability density function of a continuous type. Later we are going to call it as a normal distribution when we are discussing a standard distributions.

So, now, we will keep it as a continuous type random variable with the probability density function, 1 divided by sigma times square root of 2 pi e power minus 1 by 2 X minus μ by sigma the whole square. Always the sigma and μ values are given. One can say the μ value can lies between minus infinity to infinity whereas, the sigma quantity is always greater than 0. What is a meaning of μ and sigma? That also can be discussed.

In this example if you find out expectation of x that is minus infinity to infinity x times probability density function with the assumption that the expectation exist we will try to find the value minus infinity to infinity x times f of x dx. This is going to be after simplification you can get the answer that is μ . I am not going for the simplification of this integration as it is. If you substitute the f of x , x times f of x integration minus infinity to infinity you can get the value μ , this μ is going to be call it as a mean that is called the mean of the random variable x here.

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The image shows a screenshot of a software window titled "Random Variable_CDF_March9_2018 - Windows Journal". The window contains handwritten mathematical derivations in red ink on a lined background. The first equation is $E(x^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$, which is simplified to $= \mu^2 + \sigma^2$. The second equation is $Var(x) = E((x-\mu)^2) = E(x^2) - (E(x))^2$, which is simplified to $= \mu^2 + \sigma^2 - \mu^2 = \sigma^2$. The window also shows a taskbar at the bottom with various icons and a system tray.

Similarly, if you compute expectation X square nothing but minus infinity to infinity x square times f of x dx . One can able to get by after some simplification you can get μ square plus σ square substituting f of x is 1 divided by σ times square root of 2 pi power minus 1 by 2 X minus μ by σ whole square therefore, the variance of the random variable X that is expectation of X minus the mean is μ the whole square that is same as expectation of X square minus expectation of X the whole square.

Just now we got it E of X square is μ square plus σ square and expectation of X that is mean that we got it as a μ that is minus μ square therefore, you get variance is σ square. That means, for a for this continuous type random variable the mean is going to be μ and the variance is going to be σ square.

We have another measure that is a positive square root of variance that is called as standard deviation. So, here the σ is the standard deviation because σ square is a variance and the positive square root of variance that is called standard deviation for this continuous type random variable, the σ is the standard deviation and σ square is the variance. So, we got first moment that is the mean, variance we got it a σ square now we can go for higher order moments.

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For $n=3,4,\dots$

$$E[(X-\mu)^n] = \int_{-\infty}^{\infty} (x-\mu)^n \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$= \begin{cases} 0, & n=3,5,\dots \\ ((n-1)(n-3)\dots 3 \times 1) \sigma^n, & n=2,4,6,\dots \end{cases}$$

That is expectation of X minus μ power n for n is equal to 3 onwards, because for n is equal to 2 that is famous variance we got it already. So, we are computing the n th order moment about the mean from 3 onwards, that is same as minus infinity to infinity, x minus μ is a mean power $n-1$ divided by sigma square root of 2π e power minus 1 by 2, x minus μ by sigma the whole square dx .

If you see the integration very carefully when n is a odd positive integer then the integration value is going to be 0 because e power minus 1 by 2, x minus μ divided by sigma whole square is a positive is a even function when n is odd positive integer the whole integration values is going to be 0 therefore, you can immediately conclude, this is going to be 0 for n is equal to 3 5 and so on. Now the question is what is a value when n is going to be the even positive integer. By even positive integer one can simplify this integration and you can get the answer that is $n-1, n-3$ and so on till on to (Refer Time: 24:03) of 3 into 1 times sigma power n when n is going to be 2 4 6 and so on.

So, for this continuous type random variable which is nothing but the normal distribution with the mean μ and the variance σ^2 we are finding the n th order moment about the mean for all for all the odd powers it is going to be 0 for the even you get the expression is $n-1$ into $n-3$ and so on, 3 into 2 times sigma power n , when n takes a positive even positive integers.